

# Spreading Processes and Large Components in Ordered, Directed Random Graphs

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## Abstract

Order the vertices of a directed random graph  $v_1, \dots, v_n$ ; edge  $(v_i, v_j)$  for  $i < j$  exists independently with probability  $p$ . This random graph model is related to certain spreading processes on networks. We consider the component reachable from  $v_1$  and prove existence of a sharp threshold  $p^* = \log n/n$  at which this reachable component transitions from  $o(n)$  to  $\Omega(n)$ .

## 1 Introduction

In this note we study a random graph model that captures the dynamics of a particular type of spreading process. Consider a set of  $n$  ordered vertices  $\{v_1, \dots, v_n\}$  with vertex  $v_1$  initially ‘infiltrated’ (at time step 1). At time steps  $2, 3, \dots, n$ , vertex  $v_1$  attempts to independently infiltrate, with probability  $p$ , each of  $v_2, v_3, \dots, v_n$  in turn (one per step). Either  $v_i$  gets infiltrated or immunized. If  $v_i$  is infected, it attempts to infect  $v_{i+1}, \dots, v_n$ , also each with probability  $p$ ;  $v_i$  does not attempt to infect  $v_1, \dots, v_{i-1}$ , however, as prior vertices are already either infiltrated or immunized. At time step  $i$ , all infiltrated vertices  $v_j$  with  $j < i$  are attempting to infiltrate  $v_i$ , and  $v_i$  gets infiltrated if any one of these attempts succeeds. Intuitively,  $v_i$  is more likely to get infiltrated if more vertices are already infiltrated at the time that  $v_i$  becomes ‘susceptible’. One example of such a contagion process is given in [6].

This spreading process is equivalent to the following random model of an ordered, directed graph  $G$ : order the vertices  $v_1, \dots, v_n$ , and for  $i < j$ , the directed edge  $(v_i, v_j)$  exists in  $G$  with probability  $p$  (independently). Vertex  $v_i$  is infected if there is a (directed) path from  $v_1$  to  $v_i$ . The question we address is, “What is the size of the set of vertices reachable from  $v_1$ ?” (the size of the infection). We prove the following sharp result.

**Theorem 1.** *Let  $\mathcal{R}$  be the set of vertices reachable from  $v_1$ , and suppose  $p = \frac{c \log n}{n} + \xi(n)$ , where  $\xi(n) = o(\frac{\log n}{n})$  and  $c > 0$  is fixed. Then:*

1. *If  $c < 1$ , then  $|\mathcal{R}| = n^{c+o(1)}$ , a.a.s.*

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2. If  $c = 1$ , then  $|\mathcal{R}| = o(n)$ , a.a.s.
3. If  $c > 1$ , then  $|\mathcal{R}| = \left(1 - \frac{1}{c} + o(1)\right) n$ , a.a.s.

Recall that an event holds a.a.s. (asymptotically almost surely), if it holds with probability  $1 - o(1)$ ; that is it holds with probability tending to one as  $n$  tends to infinity. Note that we do not explicitly care whether  $\xi(n)$  is positive or negative in the results above.

Similar phase transitions are well known for various graph properties in other random graph models. As shown by Erdős and Rényi in [2], in the  $G(n, M)$  model of random graphs, where a graph is chosen independently from all graphs with  $M$  edges, there is a similar emergence of a component of size  $\Theta(n)$  around  $M = \frac{n}{2}$  edges. Likewise, a threshold for connectivity was shown for  $M = \frac{n \log n}{2}$  edges. For the more familiar  $G(n, p)$  model, where edges are present independently with probability  $p$ , this translates into a threshold at  $p = \frac{1}{n}$  for a giant component, and at  $p = \frac{\log n}{n}$  for connectivity. A much more comprehensive account of results on properties of random graphs can be found in [1]. Łuczak in [4] and more recently Łuczak and Seierstad in [5], studied the emergence of the giant component in a random directed graphs, in both the directed model where  $M$  random edges are present and in the model where edges are present with probability  $p$ . Thresholds for strong connectivity were established for random directed graphs by Palásti [7] (for random directed graphs with  $M$  edges) and Graham and Pike [3] (for random directed graphs with edge probability  $p$ ). We are not aware of any results for ordered directed random graphs where edges connect vertices of lower index to higher index.

## 2 A Proof of Theorem 1

*Upper bounds:* For  $i > 1$ , let  $\mathcal{R}_i$  denote the event that  $v_i$  is reachable, and let  $X_i$  denote the number of paths to vertex  $v_i$  in  $G$ . If  $\mathcal{P}_i$  denotes the set of all potential paths from  $v_1$  to  $v_i$ , then  $X_i = \sum_{x \in \mathcal{P}_i} I(x)$  where  $I(x)$  is a  $\{0, 1\}$  indicator random variable indicating whether the path  $x$  exists in  $G$ ;  $I(x) = 1$  if and only if all edges in the path  $x$  are present in  $G$ . Then,

$$\begin{aligned} \mathbb{P}(\mathcal{R}_i) &= \mathbb{P}(X_i \geq 1) \leq \mathbb{E}[X_i] = \sum_{x \in \mathcal{P}_i} \mathbb{E}[I(x)] \\ &= \sum_{\ell=0}^{i-2} \sum_{\substack{x \in \mathcal{P}_i \\ |x|=\ell+1}} \mathbb{E}[I(x)] \\ &= \sum_{\ell=0}^{i-2} \binom{i-2}{\ell} p^{\ell+1} = p(1+p)^{i-2} \leq pe^{pi}. \end{aligned}$$

Let  $X$  denote the number of reachable vertices (other than  $v_1$ ).

$$\mathbb{E}[X] = \sum_{i=2}^n \mathbb{P}(\mathcal{R}_i) \leq \sum_{i=1}^n pe^{pi} = p \cdot \frac{e^{p(n+1)} - 1}{e^p - 1}.$$

For  $p = \frac{c \log n}{n} + \xi(n)$  with  $c < 1$ ,

$$e^{p(n+1)} - 1 = \exp(c \log n + o(\log n)) - 1 = n^{c+o(1)},$$

and

$$\frac{p}{(e^p - 1)} = \left( \sum_{k=1}^{\infty} \frac{p^{k-1}}{k!} \right)^{-1} = 1 + O(p).$$

Thus,

$$\mathbb{E}[X] \leq n^{c+o(1)}.$$

Applying Markov's inequality yields that  $\mathbb{P}(X > \log(n)\mathbb{E}[X]) = o(1)$ , so  $X \leq \log(n)\mathbb{E}[X] = n^{c+o(1)}$ , a.a.s.

Now consider  $c > 1$ . Let

$$\begin{aligned} \xi'(n) &:= \frac{3}{c} \max \left\{ \frac{n}{\log \log n}, \frac{n^2 \xi(n)}{c \log n} \right\} \\ t &:= \frac{n}{c} - \xi'(n) \end{aligned}$$

Note that by our choice of  $\xi'(n)$ , and the fact that  $\xi(n) = o(\frac{\log n}{n})$ , that  $\xi'(n) = o(n)$ . Then,

$$\begin{aligned} \mathbb{P}(\mathcal{R}_t) &\leq pe^{pt} = p \exp \left( \left( c \frac{\log n}{n} + \xi(n) \right) \left( \frac{n}{c} - \xi'(n) \right) \right) \\ &= p \exp \left( \log(n) + \frac{n \xi(n)}{c} - \frac{c(\log n) \xi'(n)}{n} - \xi(n) \xi'(n) \right) \\ &\leq (1 + o(1))c \exp \left( \log \log(n) + \frac{n \xi(n)}{c} - \frac{c(\log n) \xi'(n)}{n} \right) = o(1) \end{aligned}$$

Here, the last inequality comes from the fact that, by our choice of  $\xi'(n)$ ,

$$\frac{c(\log(n)) \xi'(n)}{n} - \log \log(n) - \frac{n \xi(n)}{c} \geq \frac{1}{3} \xi'(n).$$

Since  $pe^{pi}$  is increasing in  $i$ , the expected number of reachable vertices  $v_i$  with  $i \leq t$  is at most  $t\mathbb{P}(\mathcal{R}_t) = o(n)$ . Applying Markov's inequality,  $|\mathcal{R} \cap \{v_1, \dots, v_t\}| = o(n)$  a.a.s. Thus,

$$|\mathcal{R}| \leq n - t + |\mathcal{R} \cap \{v_1, \dots, v_t\}| = \left(1 - \frac{1}{c} + o(1)\right) n \text{ a.a.s.}$$

For  $p = \frac{\log n}{n} + \xi(n)$  with  $\xi(n) = o\left(\frac{\log n}{n}\right)$ , we will write  $\xi(n) = \omega(n) \frac{\log n}{n}$ , where  $\omega(n) \rightarrow 0$ . Let  $t = n \cdot \left(1 - \omega(n) - \frac{1}{\log \log n}\right)$ . Then,

$$\begin{aligned} \mathbb{P}(\mathcal{R}_t) &\leq pe^{-pt} = \exp \left[ (1 + \omega(n)) \left(1 - \omega(n) - \frac{1}{\log \log n}\right) \log n + \log \left( (1 + \omega(n)) \frac{\log n}{n} \right) \right] \\ &= \exp \left[ -\omega(n)^2 \log n - (1 + \omega(n)) \frac{\log n}{\log \log n} + \log \log n + \log(1 + \omega(n)) \right] \\ &= o(1), \end{aligned}$$

Thus the expected value of  $|\mathcal{R} \cap \{v_1, \dots, v_t\}|$  is  $o(n)$  and by Markov's inequality, this is also true a.a.s. Now, since  $n - t$  is also  $o(n)$ , we have that  $R = o(n)$  a.a.s.  $\square$

To prove the lower bounds, we require a simple lemma similar to Dirichlet's theorem. Let  $d(i)$  denote the number of divisors of  $i$  and let  $d_t(i)$  denote the number of divisors of  $i$  that are at most  $t$ . Dirichlet's Theorem states that

$$\sum_{i=1}^k d(i) = k \log k + (2\gamma - 1)k + O(\sqrt{k}),$$

where  $\gamma$  is Euler's constant. For our purposes, we need a refinement of this result, summing  $d_t(i)$ .

**Lemma 1.**  $\sum_{i=1}^k d_t(i) = k \log \min(t, k) + O(k).$

*Proof.* For  $t > k$  the result follows from Dirichlet's theorem as we may replace  $d_t(i)$  with  $d(i)$  in the summation. For  $t \leq k$ ,

$$\sum_{i=1}^k d_t(i) = k + \left\lfloor \frac{k}{2} \right\rfloor + \left\lfloor \frac{k}{3} \right\rfloor + \cdots + \left\lfloor \frac{k}{t} \right\rfloor \leq k \mathcal{H}_t,$$

where  $\mathcal{H}_t$  is the  $t$ -th harmonic number. □

*Lower bounds:* For exposition, assume that we construct our graph on countably many vertices and that we then restrict our attention to the first  $n$  vertices. Let  $X_i$  denote the index of the  $i$ -th reachable vertex (that is not  $v_1$ ). If  $X_i > n$  then  $|\mathcal{R}| \leq i$ . Set  $X_0 = 1$ , and for  $i \geq 1$ ,  $X_i - X_{i-1}$  is geometrically distributed with parameter  $1 - (1 - p)^i$ . Fix  $t$ , and consider  $\mathbb{E}[X_t]$ :

$$\mathbb{E}[X_t] = \sum_{k=1}^t \mathbb{E}[X_k - X_{k-1}] = \sum_{k=1}^t \frac{1}{1 - (1 - p)^k}.$$

Each term is an infinite geometric series, and so

$$\mathbb{E}[X_t] = \sum_{k=1}^t \sum_{j=0}^{\infty} (1 - p)^{kj}.$$

As this series is absolutely summable (as  $\mathbb{E}[X_t]$  is clearly finite), Fubini's theorem allows us to rearrange terms in the summation to get

$$\mathbb{E}[X_t] = t + \sum_{k=1}^t \sum_{j=1}^{\infty} (1 - p)^{kj} = t + \sum_{i=1}^{\infty} d_t(i) (1 - p)^i.$$

because the term  $(1 - p)^i$  appears in the original summation (where  $i = kj$ ) once for every divisor  $i$  has that is at most  $t$ . We now use summation by parts to manipulate the second term:

$$\begin{aligned} \sum_{i=1}^{\infty} d_t(i) (1 - p)^i &= p \sum_{i=1}^{\infty} (1 - p)^{i-1} \left( \sum_{\ell=1}^i d_t(\ell) \right) \\ &= p \sum_{i=1}^{\infty} (1 - p)^{i-1} (i \log(\min\{t, i\}) + O(i)) \\ &\leq p(\log t + O(1)) \sum_{i=1}^{\infty} i (1 - p)^{i-1}. \end{aligned}$$

Since  $\sum_{i=1}^{\infty} i(1-p)^{i-1} = 1/p^2$ , we have that

$$\mathbb{E}[X_t] = t + \frac{\log t}{p} + O\left(\frac{1}{p}\right). \quad (1)$$

Furthermore, since  $X_{k+1} - X_k$  and  $X_k - X_{k-1}$  are independent,

$$\begin{aligned} \text{Var}(X_t) &= \sum_{k=1}^t \frac{p}{(1 - (1-p)^k)^2} \\ &\leq \sqrt{\left(\sum_{k=1}^t \frac{p^2}{(1 - (1-p)^k)^3}\right) \left(\sum_{k=1}^t \frac{1}{(1 - (1-p)^k)}\right)} \\ &\leq \sqrt{\frac{t}{p} \mathbb{E}[X_t]}. \end{aligned} \quad (2)$$

Here, the first inequality follows from an application of Cauchy-Schwarz, and the second from  $\frac{p^2}{(1-(1-p)^k)^3} \leq \frac{p^2}{p^3} = \frac{1}{p}$ .

Now, suppose that  $p = c \frac{\log n}{n} + \xi(n)$  for  $c < 1$ , and set  $t = n^c \exp(-n|\xi(n)| - \log \log(n))$ . Then, from eq:Et,

$$\mathbb{E}[X_t] \leq n^c \exp(-n|\xi(n)|) + \frac{c \log n - 2n|\xi(n)| - \log \log n}{n^{-1}(c \log n + n\xi(n))} + O\left(\frac{\log n}{n}\right) \quad (3)$$

$$\leq n^c \exp(-n|\xi(n)|) + n - \frac{n^2|\xi(n)| - \log \log n}{(c \log n + n\xi(n))} + O\left(\frac{\log n}{n}\right) \quad (4)$$

$$= n \left(1 - \frac{n|\xi(n)| + \log \log n}{(c \log n + n\xi(n))} + o\left(\frac{n|\xi(n)| + \log \log n}{\log n}\right)\right). \quad (5)$$

For  $n$  sufficiently large,  $\mathbb{E}[X_t] \leq n \left(1 - \frac{n|\xi(n)| + \log \log n}{2c \log n}\right)$ . Meanwhile, from (2),

$$\text{Var}(X_t) \leq (1 + o(1)) \sqrt{\frac{n^c}{\log n} \cdot (1 + o(1)) \frac{n}{c \log n} \cdot \mathbb{E}[X_t]} = \frac{n^{\frac{1}{2}(1+c)}}{\log n} \sqrt{\frac{\mathbb{E}[X_t]}{c}} = O\left(\frac{n^{3/2}}{\log n}\right),$$

because  $\mathbb{E}[X_t] = O(n)$  and  $c < 1$ . Chebyshev's inequality asserts that

$$\mathbb{P}\left[|X_t - \mathbb{E}[X_t]| \geq \frac{n^2|\xi(n)| + n \log \log n}{2c \log n}\right] \leq \frac{4c^2 \log^2 n \cdot \text{Var}(X_t)}{(n^2|\xi(n)| + n(\log \log n))^2} = o(1).$$

Thus,  $\mathbb{P}\left[X_t \leq \mathbb{E}[X_t] + \frac{n \log \log n}{2c \log n}\right] = 1 - o(1)$ . Using eq:Et1,

$$\mathbb{P}\left[X_t \leq n \left(1 - \frac{\log \log n}{2c \log n} + o\left(\frac{\log \log n}{c \log n}\right)\right)\right] = 1 - o(1),$$

i.e.,  $X_t < n$  a.a.s. Since  $X_t < n$  implies  $|\mathcal{R}| \geq t$ , we have that  $|\mathcal{R}| > n^c \exp(-n|\xi(n)| - \log \log(n)) = n^{c+o(1)}$  a.a.s.

For  $c > 1$ , take  $t = \frac{n \log \log n}{\log n}$ . Then, using eq:Et,

$$\mathbb{E}[X_t] \leq \frac{n}{c} + o(n).$$

Again, by (2) and because  $\mathbb{E}[X_t] = O(n)$ ,  $\text{Var}(X_t) = O(n^{3/2}\sqrt{\log \log n}/\log n) = o(n^{3/2})$ . Chebyshev's inequality asserts that

$$\mathbb{P}\left[|X_t - \mathbb{E}[X_t]| \geq n^{3/4}\right] \leq \frac{o(n^{3/2})}{n^{3/2}} = o(1).$$

Hence,

$$\mathbb{P}\left[X_t \leq \mathbb{E}[X_t] + n^{3/4}\right] = 1 - o(1). \quad (6)$$

So,  $X_t \leq \frac{n}{c} + o(n)$  a.a.s. We now consider the vertices indexed higher than  $X_t$  and show that essentially all of them are reachable. Let  $Y$  be the vertices with index higher than  $X_t$  which are *not* adjacent to one of the first  $t$  reachable vertices in  $v_1, \dots, v_{X_t}$ . Then

$$\mathbb{E}[|Y|] = \sum_{j=X_t+1}^n (1-p)^t = (n - X_t)(1-p)^t \leq ne^{-pt} = \frac{n}{\log^{c+o(1)} n} = o(n).$$

Applying Markov's inequality,  $|Y| = o(n)$  with probability  $1 - o(1)$ . Since the set of vertices indexed above  $X_t$  that is not reachable is a subset of  $Y$ ,  $|\mathcal{R}| \geq t + (n - X_t) - |Y|$ . Since  $|Y|, t$  are  $o(n)$  and  $X_t = \frac{n}{c} + o(n)$ , we have that  $|\mathcal{R}| \geq n(1 - \frac{1}{c} + o(1))$  with probability  $1 - o(1)$ , as desired.  $\square$

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